# Version Control and Beyond

Leveraging Git for ML Experiment Management

### A bit about me...

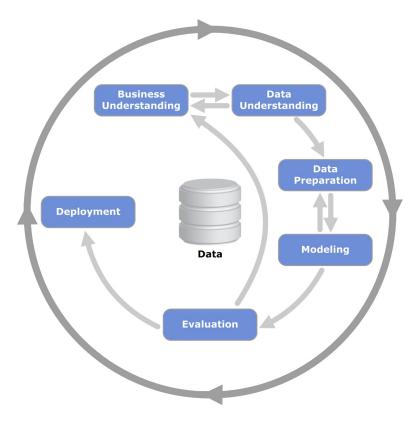
- Hi, I'm Eryk
- Sr. Data Scientist at bol ( e-commerce platform). I'm working on forecasting problems.
- I write data science articles on Medium.
- I published two editions of Python for Finance Cookbook (Packt)
- Hobbies: reading , video games



### Agenda

- 1. Why track experiments?
- 2. Why DVC?
- 3. Overview of how it works
- 4. Demo

### A typical ML workflow



### Sweet spot of ML experiments



OR



Fast experimentation

Reproducibility

### Sweet spot of ML experiments



Fast experimentation



Reproducibility

### A real-life horror story

"I remember that about 15 months ago, we ran an experiment that achieved a much better score than our current model. Can you quickly find out what we did back then?"

Made-up colleague

### Reproducibility approach #1



### Reproducibility approach #2

timestamp	model_type	max_depth	n_estimators	accuracy
2023-11-10 12:10	rf	10	10	0.56
2023-11-10 12:30	rf	15	100	0.61
2023-11-15 15:15	lgbm	10	42	0.62
2023-12-03 9:00	rf	31	100	0.55

### Reproducibility approach #2 cntd.

timestamp	model_type	git commit	model	data	max_depth	n_estimators	accuracy	recall
2023-11-10 12:10	rf	341and8	rf.joblib	dataset_v1.csv	10	10	0.56	0.4
2023-11-10 12:30	rf	129any8	rf_1.joblib	dataset_v1.csv	15	100	0.61	0.45
2023-11-15 15:15	lgbm	472hab3	lgbm.joblib	dataset_v1.csv	10	42	0.62	0.47
2023-12-03 9:00	rf	876hsk1	rf_2.joblib	dataset_final.csv	31	100	0.55	0.3

### Some of the available solutions





















### Why DVC?

- You are already ( ≤) tracking code with Git
- DVC builds on top of Git
- DVC is open-source
- Works with different storage providers (AWS, GCP, Azure, GDrive, local)
- No metric server needed
- DVC's CLI is similar to Git's
- VS Code extension



```
git init dvc init git checkout dvc checkout git add dvc add git push dvc push git pull dvc pull git fetch dvc fetch git diff
```

### Reproducible ML experiments



Code

### Reproducible ML experiments

Code

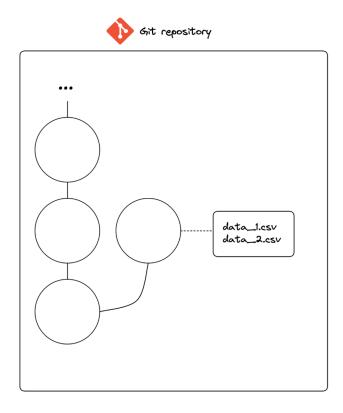


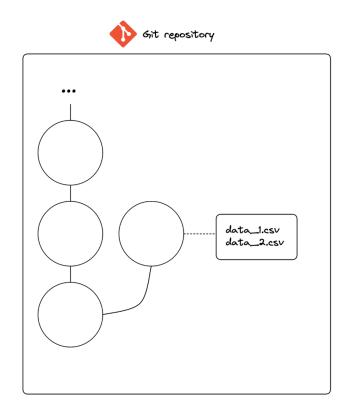
- Data
- Parameters
- Metrics
- Models
- Plots
- Other artifacts

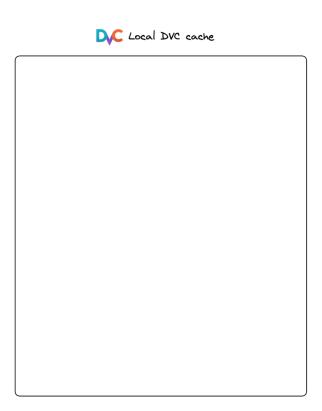
### Reproducible experimentation with DVC 101

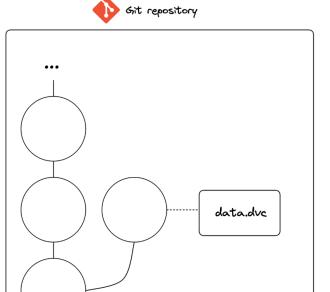
### Key components:

- 1. Versioning data and artifacts
- 2. DVC pipelines
- 3. Experiments



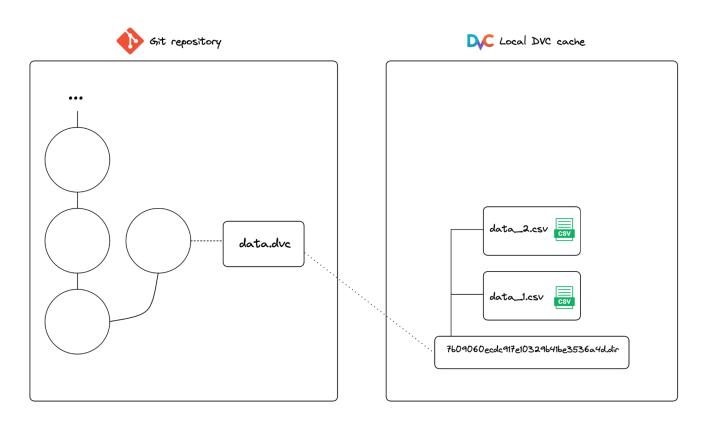


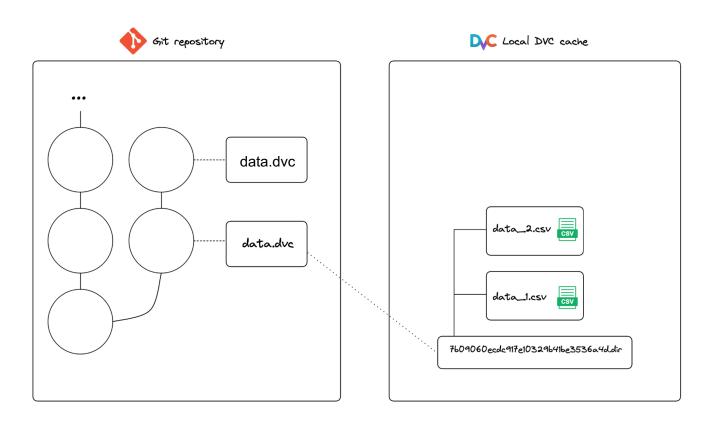


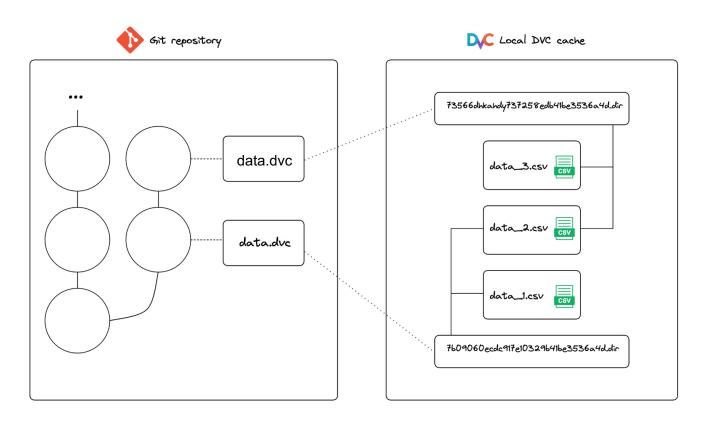


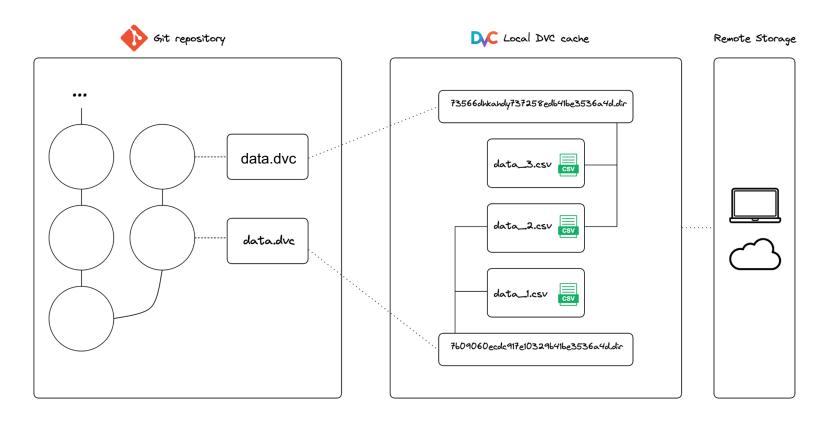
#### data.dvc











### **DVC** pipelines

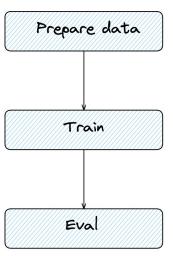
```
• • •
 cmd: python src/prepare_data.py
```

dvc.yaml

### **DVC** pipelines

```
• • •
prepare_data:
 cmd: python src/prepare_data.py
train:

  models/model.joblib
```



dvc.yaml

## Experiments with



# There's quite a lot of Pokémon

- Video games are coming out since 1996.
- The image does not even cover regional forms, mega evolutions, etc.
- Total of ~ 1200 Pokémon



### Our task: classify legendary Pokémon

	Pokedex Number	Name	Total	HP	Attack	Defense	Sp. Atk	Sp. Def	Speed	Legendary	Generation	Type 1	Type 2	Mega Evolution
0	1	Bulbasaur	318	45	49	49	65	65	45	False	1	Grass	Poison	False
1	2	Ivysaur	405	60	62	63	80	80	60	False	1	Grass	Poison	False
2	3	Venusaur	525	80	82	83	100	100	80	False	1	Grass	Poison	False
3	3	Mega Venusaur	625	80	100	123	122	120	80	False	1	Grass	Poison	True
4	4	Charmander	309	39	52	43	60	50	65	False	1	Fire	NaN	False
1190	1006	Iron Valiant	590	74	130	90	120	60	116	False	9	Fairy	Fighting	False
1191	1007	Koraidon	670	100	135	115	85	100	135	True	9	Fighting	Dragon	False
1192	1008	Miraidon	670	100	85	100	135	115	135	True	9	Electric	Dragon	False
1193	1009	Walking Wake	590	99	83	91	125	83	109	True	9	Water	Dragon	False
1194	1010	Iron Leaves	590	90	130	88	70	108	104	True	9	Grass	Psychic	False



Legendary: True



Legendary: False

Source: https://pokemondb.net/

## Demo time

### Wrapping up

- Git + DVC + VS Code = One-stop shop for ML experimentation
  - Git handles versioning for code, configuration, and small text files (metadata/pointers).
  - DVC facilitates the creation of ML pipelines and experiments, while also managing the versioning of artifacts such as data, models, plots, metrics, etc.
  - VS Code serves as a user-friendly interface for experiment management and visualizations.

### References

#### Docs:

https://dvc.org/doc

My articles on setting up experimentation with DVC:

- https://towardsdatascience.com/turn-vs-code-into-a-one-stop-shop-for-ml-experiments-49c97c47db27
- <a href="https://towardsdatascience.com/enhance-your-ml-experimentation-workflow-with-real-time-plots-434106b1a1c2">https://towardsdatascience.com/enhance-your-ml-experimentation-workflow-with-real-time-plots-434106b1a1c2</a>
- <a href="https://towardsdatascience.com/the-minimalists-guide-to-experiment-tracking-with-dvc-f07e4636bdbb">https://towardsdatascience.com/the-minimalists-guide-to-experiment-tracking-with-dvc-f07e4636bdbb</a>

A comparison of experiment tracking tools:

• <a href="https://medium.com/towards-data-science/a-comprehensive-comparison-of-ml-experiment-tracking-tools-9f0192543feb">https://medium.com/towards-data-science/a-comprehensive-comparison-of-ml-experiment-tracking-tools-9f0192543feb</a>

How DVC uses git references for experiment tracking:

https://iterative.ai/blog/experiment-refs

### Thanks for listening!

### Let's stay in touch:

- linkedin.com/in/eryklewinson
- medium.com/@eryk-lewinson
- https://github.com/erykml

GitHub repo of this project:



Python for Finance Cookbook:

